

METHOD AND APPARATUS FOR ASSAYING A DRUG CANDIDATE TO ESTIMATE A PHARMACOKINETIC PARAMETER ASSOCIATED THEREWITH

TECHNICAL FIELD

This invention is generally directed to a method and apparatus for
5 assaying a drug candidate and, more specifically, to a method for measuring the binding
interaction between a drug candidate and sensing surface-bound biomolecules of a
biosensor to determine a binding interaction parameter of the drug candidate, and then
comparing the binding interaction parameter against a predetermined drug correlation
graph (e.g., a mathematical expression) to estimate at least one pharmacokinetic
10 parameter.

BACKGROUND OF THE INVENTION

A variety of experimental techniques are currently used to determine
chemical, physical and biological properties associated with low molecular weight
substances, particularly in the context of drug discovery. For example, researchers are
15 often concerned with determining a variety of chemical, physical and biological
properties associated with drug candidates for screening purposes. The determination
of such properties often plays a pivotal role in the drug development and screening
process.

More specifically, it has long been recognized that the intensity and
20 duration of the pharmacological effect of a systemically acting drug are functions not
only of the intrinsic activity of the drug, but also of its absorption, distribution,
metabolism, and excretion (ADME) characteristics within the human body. These so-
called ADME characteristics are all intimately related to the scientific discipline known
as "pharmacokinetics." Pharmacokinetics is commonly referred to as the study of the
25 time courses (*i.e.*, kinetics) associated with the dynamic processes of ADME of a drug
and/or its metabolites within a living organism, and is closely interrelated with the
fields of biopharmaceutics, pharmacology, and therapeutics.

Because the body delays the transport of drug molecules across membranes, dilutes them into various compartments of distribution, transforms them into metabolites, and eventually excretes them, it is often difficult to accurately predict the pharmacological effect of promising new drug candidates. Researchers, however,
5 commonly use pharmacokinetic ADME studies as one method to predict the efficacy of a drug at a site of action within the body.

Traditionally, researchers involved with preclinical ADME studies have used pharmacokinetic/mathematical models coupled with actual drug concentration data from blood (or serum or plasma) and/or urine, as well as concentration data from
10 various tissues, to characterize the behavior and “fate” of a drug within living organisms. As is appreciated by those skilled in the art, the mathematical equations associated with pharmacokinetics are generally based on models that conceive the body as a multicompartmental organism. In such models it is presumed that the drug and/or its metabolites are equitably dispersed in one or several fluids/tissues of the organism.
15 Any conglomerate of fluid or tissue which acts as if it is kinetically homogeneous may be termed a “compartment.” Each compartment acts as an isotropic fluid in which the molecules of drug that enter are homogeneously dispersed and where kinetic dependencies of the dynamic pharmacokinetic processes may be formulated as functions of the amounts or concentrations of drug and metabolites therein. Stated
20 somewhat differently, the conceptual compartments of the body are separated by barriers that prevent the free diffusion of drug among them; the barriers are kinetically definable in that the rate of transport of drug or metabolite across membrane barriers between compartments is a function of, for example, the amounts or concentrations of drug and metabolites in the compartments, the permeability of various membranes,
25 and/or the amount of plasma protein binding and general tissue binding.

More specifically, pharmacokinetic/mathematical models are commonly used by pharmacokineticists to represent drug absorption, distribution, metabolism, and excretion as functions of time within the various tissues and organs of the body. In such models, the movement of the administered drug throughout the body is concisely
30 described in mathematical terms (*e.g.*, a set of differential equations). The predictive

capability of such models lies in the proper selection and development of mathematical functions that parameterize the essential factors governing the kinetic process under consideration.

For example, a drug that is administered by intravenous injection may be assumed to distribute rapidly in the bloodstream. A pharmacokinetic/mathematical model that describes this situation may be a tank containing a volume of fluid that is rapidly equilibrated with the drug. Because a fraction of the drug in the body is continually eliminated as a function of time (*e.g.*, excreted by the kidneys and metabolized by the liver), the concentration of the drug in the hypothetical tank may be characterized by two parameters: (1) the volume of fluid in the tank that will dilute the drug, and (2) the elimination rate of the drug per unit of time, both of which are generally considered to be constant. Thus, if a known set of drug concentrations in the tank is determined at various time intervals by, for example, sampling, then the volume of fluid in the tank and rate of drug elimination may be estimated. This information may then, in turn, be used for predicting the disposition of the drug within a human body.

Theoretically, an unlimited number of models may be constructed to describe the kinetic processes of drug absorption, distribution, metabolism, and excretion within the various tissues and organs of the human body. In general, however, the number of useful models is limited due to practical considerations associated with blood, tissue and/or organ sampling. As a result, and as is appreciated by those skilled in the art, two major types of models have been developed by pharmacokineticists: (1) compartmental models; and (2) physiologic models.

In pharmacokinetic compartmental models, the body is represented as a series of compartments that communicate reversibly with each other. Each compartment is not a real physiological or anatomic region; rather, each compartment is considered to be inclusive of all tissues that have similar blood flow and drug affinity. For example, a compartmental model may consist of one or more peripheral compartments representing tissue(s) connected to a central compartment representing the blood stream. Conceptually, the drug moves dynamically into and out of the central

compartment and into and out of each of the peripheral compartments. As such, rate constants may be used to represent the overall rate process for the drug's disposition within each compartment. Compartment models are generally based on linear assumptions using linear differential equations, and are particularly useful when there is
5 little information known about the tissues and their respective drug concentrations.

In contrast, pharmacokinetic physiologic models are based on known anatomic and physiologic data, data which is kinetically described in view of the actual blood flow volumes responsible for distributing the drug to the various parts of the body. Because there are many tissue organs in the body, each tissue volume must be
10 estimated and its drug concentration and rate of change described mathematically (tissues having similar blood perfusion properties, however, are typically grouped together). Unfortunately, much of the information required to adequately describe such pharmacokinetic physiologic models are often very difficult to obtain experimentally. Nevertheless, such physiologically based models are commonly used in conjunction
15 with animal data and interspecies scaling techniques to predict the drug's disposition within a human body.

More importantly, however, is that pharmacokinetic/mathematical models, and knowledge of their associated ADME parameters play an extremely important role in drug discovery and development. A typical example is a drug that is
20 active following intravenous administration but is considerably less active after comparable oral doses. Having appropriate pharmacokinetic information may reveal (1) whether the drug was poorly absorbed to yield subtherapeutic circulating levels, or (2) whether the drug experienced presystemic metabolism to an inactive metabolite. Such information may also provide guidance for subsequent decisions, such as (1) whether to
25 improve drug absorption by altering the salt form or formulation, (2) whether to investigate the possibility of making prodrugs, or (3) whether to consider a parenteral route of administration.

In addition to the foregoing, pharmacokinetic/mathematical models are also generally considered extremely useful for, among other things: (1) predicting
30 plasma, tissue, and urine drug levels with any dosage regimen; (2) calculating the

optimum dosage regimen for an individual patient; (3) estimating the possible accumulation of drugs and/or metabolites; (4) correlating drug concentrations with pharmacologic and toxicologic activity (*i.e.*, pharmacodynamics); (5) evaluating differences in the rate or extent of availability between formulations (*i.e.*, bioequivalence); (6) describing how changes in physiology or disease affect the absorption, distribution, and/or elimination of the drug; and (7) explaining drug-drug and food-drug interactions.

Lastly, pharmacokinetic ADME data has also become an integral part of the pharmacological characterization process of promising new drug candidates. Regulatory agencies, such as the U.S. Food and Drug Administration, now require (1) a determination of pharmacokinetic ADME data in Phase I drug studies, and (2) a submission of pharmacokinetic ADME data as part of a New Drug Application. In this context, such pharmacokinetic ADME data is deemed essential for predicting the behavior and fate of the drug candidate within the human body.

Accordingly, there is a need in the art for improved methods for determining one or more pharmacokinetic parameters associated with absorption, distribution, metabolism, and excretion of a drug candidate. There is also a need for apparatuses useful for carrying out such methods. The present invention fulfills these needs and provides further related advantages.

SUMMARY OF THE INVENTION

In brief, the present invention is directed to a method and apparatus for assaying a drug candidate. More specifically, this invention discloses a method for measuring the binding interaction between a drug candidate and sensing surface-bound biomolecules of a biosensor to determine a binding interaction parameter of the drug candidate, and then comparing the binding interaction parameter against a predetermined drug correlation graph to estimate at least one pharmacokinetic parameter of the drug candidate. The at least one pharmacokinetic parameter may, for example, be one or more of ADME.

In another embodiment of the present invention, at least two pharmacokinetic parameters of the drug candidate are determined, and in yet another embodiment, at least one pharmacokinetic parameter and a solubility property of the drug candidate are determined. Such pharmacokinetic parameters and/or solubility property may be determined when the one or more sensing surface-bound biomolecules are selected from, for example, liposomes, plasma proteins, CYP 450 enzymes, metabolic enzymes, or transport proteins.

The biosensor used in the practice of the present invention may utilize a mass-sensing technique, such as surface plasmon resonance. In addition, the biosensor may further employ a sensor chip, wherein the sensor chip comprises a free electron metal that includes a sensor surface, wherein the free electron metal is copper, silver, aluminum or gold. The sensor chip may further comprise a hydrogel coupled to the sensor surface, wherein the hydrogel has a plurality of functional groups, and wherein the one or more sensing surface-bound biomolecules are covalently bonded to the hydrogel.

In a more specific embodiment, a sensor surface adopted for use with a biosensor is disclosed. The sensor surface comprises a hydrogel matrix coating coupled to a top surface of the sensor surface, wherein the hydrogel matrix coating has plurality of functional groups. At least two different liposomes are bonded to the plurality of functional groups at discrete and noncontiguous locations on the hydrogel mixtrix coating of the sensor surface. In one embodiment, the sensor surface is a sensor chip, and a free electron metal is interposed between the hydrogel matrix and the top surface of the sensor surface.

In another embodiment of this invention, an apparatus is disclosed for assaying a drug candidate, wherein the apparatus comprises a biosensor having one or more sensing surface-bound biomolecules associated therewith and capable of measuring at least one binding interaction parameter of the drug candidate, and a computer memory containing a data structure for comparing the at least one binding interaction parameter against at least one mathematical expression correlated from

binding interaction data associated with known drug compounds to determine an estimate of at least one pharmacokinetic parameter of the drug candidate.

In yet a further embodiment, a computer memory containing a data structure useful for assaying a drug candidate in accordance with the methods of the present invention is disclosed (as well as a generated data signal conveying the same). The data structure may be used to determine an estimate of at least pharmacokinetic parameter of the drug candidate.

These and other aspects of the present invention will be evident upon reference to the following detailed description and related Figures.

10 BRIEF DESCRIPTION OF THE DRAWINGS

Figure 1A illustrates a sensorgram reflective of the steady-state binding levels associated with a selected drug-biomolecule interaction.

Figure 1B illustrates an enlarged portion of the sensorgram of Figure 1A and shows sensorgram irregularities.

15 Figure 1C illustrates an enlarged portion of the sensorgram of Figure 1A, wherein bulk-refractive index effects have been eliminated.

Figure 2A illustrates a single "dip" depicting the reflected light intensity associated with a homogeneous sensor surface.

20 Figure 2B illustrates a number of "dips" depicting the non-averaged reflected light intensities associated with a non-homogeneous sensor surface.

Figure 2C illustrates a broadening of the "dip" depicting the averaged reflected light intensities associated with a non-homogeneous sensor surface.

Figure 3 shows a high-level block diagram of an exemplary computer system for assaying a drug candidate in accordance with the methods of the present invention.

Figure 4 depicts a correlation graph having known K_D values for known compounds plotted along the abscissa (*i.e.*, the x-axis) and corresponding measured K_D values obtained via the biosensor plotted along the ordinate (*i.e.*, the y-axis) for nine drugs with known levels of plasma protein binding

Figure 5 depicts a correlation graph having respective binding levels at 100 μM (R 100 μM) divided by the molecular weight of known drug compounds' plotted along the abscissa (*i.e.*, the x-axis) and corresponding human serum albumin binding percentage, as measured by equilibrium dialysis, plotted along the ordinate (*i.e.*, the y-axis).

Figure 6 depicts reference subtracted sensorgram traces for each of three drug candidates A-C.

Figure 7 depicts a correlation graph having known fraction absorbed in humans (FA%) plotted along the ordinate (*i.e.*, the y-axis) and corresponding calibrated (*i.e.*, reference subtracted) steady state binding levels for each drug at 500 μM plotted along the abscissa in a 10-logarithm scale (*i.e.*, the x-axis).

DETAILED DESCRIPTION OF THE INVENTION

The present invention is directed to a method for assaying a drug candidate and, more specifically, to a method and apparatus for measuring the binding interaction between at least one drug candidate and sensing surface-bound biomolecules of a biosensor to determine a binding interaction parameter of the drug candidate, and then comparing the binding interaction parameter against a predetermined drug correlation graph (*e.g.*, a mathematical expression fitted to a series of known data points) to determine an estimate of at least one pharmacokinetic parameter. Although many specific details of certain embodiments of the invention are set forth in the following detailed description and accompanying Figures, those skilled in the art will recognize that the present invention may have additional embodiments, or that the invention may be practiced without several of the details described herein.

For purposes of clarity and to assist in understanding the full scope of the present invention, a brief review of the nomenclature associated with pharmacokinetics has been provided. As used within the context of the present invention, the following pharmacokinetic terms shall be construed broadly, and shall have their generally accepted meanings as set forth below. (As previously noted, "pharmacokinetics" refers to the study of the kinetics associated with the dynamic processes of absorption,

distribution, metabolism, and excretion (ADME) of a drug and/or its metabolites within a living organism.)

“Absorption” refers to the process of uptake of a drug compound from the site of administration into the systemic circulation. The transfer of drug across the intestinal lumen is generally referred to as oral absorption, whereas the transfer of drug across an external physiological barrier is referred to general absorption. As disclosed herein, the pharmacokinetic parameter of absorption may be estimated from biosensor data associated with a sensor chip having, for example, a plurality of appropriate liposomes immobilized thereon.

“Distribution” refers to the transfer of a drug compound from the site of administration to the total systemic circulation and then to extracellular and intracellular water and tissues. Drug distribution is usually a rapid and reversible process. As disclosed herein, the pharmacokinetic parameter of distribution may be estimated from biosensor data associated with a sensor chip having, for example, a plurality of appropriate plasma proteins, liposomes, and/or transport proteins immobilized thereon.

“Metabolism” refers to the sum of all the chemical reactions for biotransformation of endogenous and exogenous substances which take place in the living cell. As disclosed herein, the pharmacokinetic parameter of metabolism may be estimated from biosensor data associated with a sensor chip having, for example, a plurality of appropriate metabolic enzymes immobilized thereon.

“Excretion” refers to the final elimination or loss of a drug from the body. Drug excretion includes both passive diffusion and relative specific carrier mediated excretion. Drugs may be excreted, unchanged or as metabolites, in urine via the kidneys or in feces via the bile and/or the intestine. Volatile compounds are often excreted in expired air by the lungs. As disclosed herein, the pharmacokinetic parameter of excretion may be estimated from biosensor data associated with a sensor chip having immobilized thereon, for example, an antibody that specifically detects the drug, as well as other proteins/receptors having a high affinity/specificity against the drug candidate. Such antibodies and proteins/receptors may be used to quantify the

concentration/amount of the drug in different body fluids (*e.g.*, urine/feces) and tissues, using a direct binding assay.

In addition to these ADME parameters, the solubility of a drug is also an important property that may be measured by the methods of the present invention.

5 “Solubility” refers to the ability of two substances to form a homogeneous solution or mixture with each other. Solubility is important for the dissolution of drug given in solid dosage form. As disclosed herein, the solubility of a drug candidate may be estimated from sensorgram irregularities associated with reflectance minimum and dip-shape of biosensor data.

10 Furthermore, and as used herein, the term “parameter” refers to a constant or variable term in a function (*e.g.*, a mathematical expression) that determines the specific form of the function but not necessarily its general nature. For example, the constant term “*a*” in the function $f(x) = ax$, where “*a*” determines only the slope of the line described by $f(x)$, is referred to as a parameter. As such, the term “binding
15 interaction parameter” refers to those constant or variable terms that are related to the binding interaction between a drug candidate and a sensing surface-bound biomolecule and includes, for example, association and dissociation rate constants, as well as maximum binding capacity. Similarly, the term “pharmacokinetic parameter” refers to those constant and variable terms that are related to the disposition of the drug candidate
20 within a living organism and includes, for example: volume of distribution; total clearance; protein binding; tissue binding; metabolic clearance; renal clearance; hepatic clearance; biliary clearance; intestinal absorption; bioavailability; relative bioavailability; intrinsic clearance; mean residence time; maximum rate of metabolism; Michaelis-Menten constant; partitioning coefficients between tissues and blood (or
25 plasma) such as those partitioning coefficients associated with the blood brain barrier, blood placenta barrier, blood human milk partitioning, blood adipose tissue partitioning, and blood muscle partitioning; fraction excreted unchanged in urine; fraction of drug systemically converted to metabolites; elimination rate constant; half-life; and secretion clearance.

The methods of the present invention are intended to be carried out by use of an affinity-based biosensor. As is appreciated by those skilled in the art, "biosensors" are analytical devices for analyzing minute quantities of sample solution having an analyte of interest, wherein the analyte is analyzed by a detection device that may employ a variety of detection methods. Typically, such methods include, but are not limited to, mass detection methods, such as piezoelectric, optical, thermo-optical and surface acoustic wave (SAW) device methods, and electrochemical methods, such as potentiometric, conductometric, amperometric and capacitance methods. With regard to optical detection methods, representative methods include those that detect mass surface concentration, such as reflection-optical methods, including both internal and external reflection methods, angle, wavelength or phase resolved, for example, ellipsometry and evanescent wave spectroscopy (EWS), the latter including surface plasmon resonance (SPR) spectroscopy, Brewster angle refractometry, critical angle refractometry, frustrated total reflection (FTR), evanescent wave ellipsometry, scattered total internal reflection (STIR), optical wave guide sensors, evanescent wave-based imaging such as critical angle resolved imaging, Brewster angle resolved imaging, SPR angle resolved imaging, and the like. Further, photometric methods based on, for example, evanescent fluorescence (TIRF) and phosphorescence may also be employed, as well as waveguide interferometers.

In the detailed description and Examples that follow, the present invention is illustrated in the context of SPR spectroscopy. However, it is to be understood that the present invention is not limited to this detection method. Rather, any affinity-based detection method where an analyte binds to a ligand immobilized on a sensing surface may be employed, provided that a change in a property of the sensing surface is measured and quantitatively indicative of binding of the analyte to the immobilized ligand thereon. In the context of SPR spectroscopy, one exemplary class of biosensors is sold by Biacore AB (Uppsala, Sweden) under the tradename BIACORE® (hereinafter referred to as "the BIACORE instrument"). Such biosensors utilize a SPR based mass-sensing technique to provide a "real-time" binding interaction analysis between a surface bound ligand and an analyte of interest.

The BIACORE instrument includes a light emitting diode, a sensor chip covered with a thin gold film, an integrated fluid cartridge and a photo detector. Incoming light from the diode is reflected in the gold film and detected by the photo detector. At a certain angle of incidence ("the SPR angle"), a surface plasmon wave is set up in the gold layer, which is detected as an intensity loss or "dip" in the reflected light. More particularly, and as is appreciated by those skilled in the art, the phenomenon of surface plasmon resonance (SPR) associated with the BIACORE instrument is dependent on the resonant coupling of light, incident on a thin metal film, to oscillations of the conducting electrons, called plasmons, at the metal film surface.

These oscillations give rise to an evanescent field which extends from the surface into the sample solution. When resonance occurs, the reflected light intensity drops at a sharply defined angle of incidence, the SPR angle, which is dependent on the refractive index within the reach of the evanescent field in the proximity of the metal surface.

Stated somewhat differently, surface plasmon resonance is an optical phenomenon arising in connection with total internal reflection of light at a metal film-liquid interface. Normally, light traveling through an optically denser medium, *e.g.*, a glass prism, is totally reflected back into the prism when reaching an interface of an optically less dense medium, *e.g.*, a buffer, provided that the angle of incidence is larger than the critical angle. This is known as total internal reflection. Although the light is totally reflected, a component of the incident light momentum called the evanescent wave penetrates a distance of the order of one wavelength into the buffer. The evanescent wave may be used to excite molecules close to the interface. If the light is monochromatic and *p*-polarized, and the interface between the media is coated with a thin (a fraction of the light wave-length) metal film, the evanescent wave under certain conditions will interact with free oscillating electrons (plasmons) in the metal film surface. When surface plasmon resonance occurs, light energy is lost to the metal film and the reflected light intensity is thus decreased.

The resonance phenomenon will only occur for light incident at a sharply defined angle which, when all else is kept constant, is dependent on the refractive index in the flowing buffer close to the surface. Changes in the refractive index out to about

1 μm from the metal film surface can thus be followed by continuous monitoring of the
resonance angle. A detection volume is defined by the size of the illuminated area at
the interface and the penetration depth of the evanescent field. It should be noted that
no light passes through the detection volume (the optical device on one side of the metal
5 film detects changes in the refractive index in the medium on the other side).

As noted above, the SPR angle depends on the refractive index of the
medium close to the gold layer. In the BIACORE instrument, dextran is typically
coupled to the gold surface, with the ligand being bound to the surface of the dextran
layer. (Note a detailed discussion of matrix coatings for biosensor sensing surfaces is
10 provided in U.S. Patent No. 5,436,161, which is incorporated herein by reference in its
entirety.) The analyte of interest is injected in solution form onto the sensor surface
through the fluid cartridge. Because the refractive index in the proximity of the gold
film depends upon (1) the refractive index of the solution (which is constant) and, (2)
the amount of material bound to the surface, the binding interaction between the bound
15 ligand and analyte can be monitored as a function of the change in SPR angle.

A typical output from the BIACORE instrument is a "sensorgram,"
which is a plot of response (measured in "resonance units" or "RU") as a function of
time. An increase of 1,000 RU corresponds to an increase of mass on the sensor surface
of approximately 1 ng/mm^2 . As a sample containing the analyte contacts the sensor
20 surface, the ligand bound to the sensor surface interacts with the analyte in a step
referred to as "association." This step is indicated on the sensorgram by an increase in
RU as the sample is initially brought into contact with the sensor surface. Conversely,
"dissociation" normally occurs when sample flow is replaced by, for example, a buffer
flow. This step is indicted on the sensorgram by a drop in RU over time as analyte
25 dissociates from the surface-bound ligand. A detailed discussion of the technical
aspects of the BIACORE instrument and the phenomenon of SPR may be found in U.S.
Patent No. 5,313,264, which is incorporated herein by reference in its entirety.

In addition, a detailed discussion of the technical aspects of the biosensor
sensor chips used in connection with the BIACORE instrument may be found in U.S.
30 Patent No. 5,492,840, which is incorporated herein by reference in its entirety. This

patent discloses, among other things, that each sensor chip may have a plurality of sensing surfaces, and that such sensing surfaces may be arranged in series or parallel with respect to the fluid sample pathway of the fluid cartridge. This patent also discloses that each of the plurality of sensing surfaces of a single sensor chip may have
 5 bound thereto a unique type of ligand that is capable of interacting with an analyte in its own characteristic way.

For example, and as disclosed herein, each of the four discrete sensing surfaces of the BIACORE instrument may have immobilized thereon biomolecules such as liposomes, plasma proteins, CYP 450 enzymes, other metabolic enzymes, and/or
 10 transport/efflux proteins. By immobilizing one or a selected combination of at least two different liposomes, plasma proteins, CYP 450 enzymes, other metabolic enzymes, and/or transport/efflux proteins, onto the one or more discrete sensing surfaces of a sensor chip, one or more pharmacokinetic parameters associated with a drug candidate may be readily determined. More specifically, and in one embodiment of the present
 15 invention, it has been discovered that (1) an estimate of an "absorption" parameter of a drug candidate may be determined from the binding interactions between the drug candidate and one or more appropriate liposomes; (2) an estimate of a "distribution" parameter of a drug candidate may be determined from the binding interactions between the drug candidate and an appropriate set of plasma proteins (*e.g.*, specific and
 20 nonspecific tissue binding and tissue permeability); (3) an estimate of a "metabolism" parameter of a drug candidate may be determined from the binding interactions between the drug candidate and an appropriate set of metabolic enzymes; and (4) an estimate of an "excretion" parameter of a drug candidate may be determined from the binding interactions between the drug candidate and an appropriate set of transport proteins.

25 Within the context of the present invention, suitable liposomes for estimating an "absorption" parameter of a drug candidate include organic compounds originating from natural or synthetic lipid molecules such as glycerophospholipids, glyceroglycolipids, sphingophospholipids and sphingoglycolipids, and from the classes phosphatidyl choline, phosphatidyl etanolamine, phosphatidyl serine, phosphatidyl
 30 glycerol, phosphatidyl acid, phosphatidyl inositol, galactopyranoside,

digalactopyranoside, ceramide-phosphatidyl choline, ceramide-phosphatidyl ethanolamine, ceramide-phosphatidyl serine, ceramide-phosphatidyl glycerol, ceramide-phosphatidyl acid, ceramide-phosphatidyl inositol, sphingomyelin molecules, glucosylceramides, glucocerebrosides, galactoceramides, galactocerebrosides, gangliosides, monoacyl phosphatidyl choline, cardiolipin molecules, that may be linked to saturated or unsaturated fatty or fluorocarbon chains ranging from eight to twenty-four carbons in length where fatty chains attached to the head group can be the same or of different structure, cholesterol, lanosterol, ergosterol, stigmasterol, sitosterol and derivatives thereof capable of being incorporated into lipid membranes, N,N-dimethyl-N-octadecyl-1-octadecan ammonium chloride or bromide, (N-[1-(2,3 - dioleoyloxy)propyl]-N,N,N-trimethylammonium chloride, diacetyl phosphate, N-[2,3-dihexadecyloxy)prop-1-yl]-N,N,N-trimethylammonium chloride, bolaamphiphiles, polyglycerolmonoalkylethers, polyethoxymonoalkylethers, as well as liposome-forming molecules from the classes amphiphilic polymers, amino acids, crown ether compounds and di(acyloxy)dialkylsilanes. The liposomes of the present invention and the like may be immobilized onto the sensing surface by the technique disclosed in Example 1.

Suitable plasma proteins for estimating a "distribution" parameter of a drug candidate include proteins such as Immunoglobulin G (7s-y-globulin), IgG; Immunoglobulin A, IgA; Secretory IgA, s IgA; Immunoglobulin M (19s-y-globulin) IgM; Immunoglobulin D, IgD; Immunoglobulin E, IgE; α 1-Antitrypsin, α 1 PI, α 1 A; α 1-Antichymotrypsin, (α 1X-Glycoprotein) α 1 X; Inter- α -Trypsin inhibitor, α 1TI; Antithrombin III, (heparin cofactor) AT III; α -Thiol Proteinase inhibitor (LMW kininogen) α TPI; C1-Inactivator, C1 esterase inhibitor (α 2-neuraminoglycoprotein) C1 INA; α 2-Macroglobulin, α 2 M; α 2-Antiplasmin, α 2AP; Cystatin C (Post-y-globulin; y-Trace protein); C1q (11S protein); C1r; C1s; C2; C3 (β 1C-globulin), C4 (β 1E-globulin); C5 (β 1F-globulin); C6; C7; C8; C9; Factor B, (C3-proactivator; β 2-glycoprotein II; glycine-rich β -glycoprotein); Factor D (C3-proactivator convertase); Properdin, P; Factor I, (C3b inactivator); C4-binding Protein; Fibrinogen, FI, Prothrombin, F II; Factor V (proaccelerin); FV; Factor VII, (proconvertin), F VII; Factor VIII: C (antihemophilic factor) F VIII: C; Factor VIII - Related Antigen; FV III:

Rag (von Willebrand factor) (VWF); Factor IX (Christmas factor) F IX; Factor X, (Stuart-Power factor) F X; Factor XI (plasma thromboplastin antecedent) F XI; Factor XII (Hageman factor) F XII; Factor XIII (fibrin stabilizing factor) F XIII; high-molecular-weight (HMW) Kininogen (Fitzgerald factor); Prekallikrein (Fletcher factor);

5 Plasminogen; Protein C; Protein S; Albumin, ALB; Haptoglobin, HP, Hp 1-1, Hp 2-1, Hp 2-2; Prealbumin (transthyretin, thyroxine-binding prealbumin); Retinol-binding Protein RBP; Thyroxine-binding Globulin TBG; Transcortin (corticosteroid-binding globulin) CBG; Sex hormone-binding globulin (Steroid-binding β Globulin) SHBG; Vitamin D-binding Protein (Gc-globulin, group specific component) VDBP;

10 Transcobalamin I, TC I; Transcobalamin II, TC II; Transferrin (siderophilin) TF; Ferritin; Hemopexin, HPX; Apolipoprotein A, Apo A-1, Apo A-II; Apolipoprotein B, Apo B-48, Apo B-100; Apolipoprotein C; Apo C-I, Apo C-II, Apo C-III; Apolipoprotein E, Apo E; Apolipoprotein (a), apo (a); Serum Amyloid A, SAA; α -Fetoprotein, AFP; α 1-Acid Glycoprotein (orosomucoid) α 1 AG; Ceruloplasmin, CP;

15 Serum Amyloid P protein (9.5S α 1-glycoprotein; α 1-macroglobulin) SAP; α 2-HS Glycoprotein, α 2 HS; Fibronectin (cold insoluble globulin) FN, C-reactive Protein, CRP; β 2-microglobulin, β 2 M; Pregnancy-specific β 1-glycoprotein, SP1; α 1-microglobulin, α 1 M. The plasma proteins of the present invention and the like may be immobilized onto the sensing surface through the use of known immobilization

20 techniques as is appreciated by those skilled in the art.

Suitable metabolic enzymes for estimating a "metabolism" parameter and/or an "excretion" parameter of a drug candidate include cytochrome P450-enzymes selected from the class of cytochrome P450 enzymes having an active site characterized by a protoporphyrin IX-iron complex with thiolate from a cysteine of the enzyme

25 serving as the fifth ligand to iron. Suitable CYP 450 enzymes include CYTOCHROME P450, CYP1A1, CYP1A2, CYP2A1, 2A2, 2A3, 2A4, 2A5, 2A6, CYP2B1, 2B2, 2B3, 2B4, 2B5, 2B6, CYP2C1, 2C2, 2C3, 2C4, 2C5, 2C6, 2C7, 2C8, 2C9, 2C10, 2C11, 2C12, CYP2D1, 2D2, 2D3, 2D4, 2D5, 2D6, CYP2E1, CYP3A1, 3A2, 3A3, 3A4, 3A5, 3A7, CYP4A1, 4A2, 4A3, 4A4, CYP4A11, CYP P450 (TXAS), CYP P450 11A

30 (P450_{sc}), CYP P450 17(P450_{17a}), CYP P450 19 (P450_{arom}), CYP P450 51

- (P45014a), CYP P450 105A1, CYP P450 105B1. Other important metabolic enzymes include Glutathione-thioethers, Leukotriene C₄, butyrylcholinesterase, human serum paraoxonase/arylesterase, N-Acetyltransferase, UDP-glucuronosyltransferase (UDPGT) isoenzymes, TL PST, TS PST, drug glucosidation conjugation enzyme, the glutathione-
- 5 S-transferases (GSTs) (RX:glutathione-R-transferase), GST1, GST2, GST3, GST4, GST5, GST6, alcohol dehydrogenase (ADH), ADH I, ADH II, ADH III, aldehyde dehydrogenase (ALDH), cytosolic (ALDH1), mitochondrial (ALDH2), monoamine oxidase, MAO: Ec 1.4.3.4, MAOA, MAOB, flavin-containing monoamine oxidase, enzyme superoxide dismutase (SOD), Catalase, amidases, N1,-monogluthionyl
 - 10 spermidine, N1,N8-bis(glutathionyl) spermidine, Thioesters, GS-SG, GS-S-cysteine, GS-S-cysteinylglycine, GS-S-O₃H, GS-S-CoA, GS-S-proteins, S-carbonic anhydrase III, S-actin, Mercaptides, GS-Cu(I), GS-Cu(II)-SG, GS-SeH, GS-Se-SG, GS-Zn-R, GS-Cr-R, Cholin esterase, lysosomal carboxypeptidase, Calpains, Retinol dehydrogenase, Retinyl reductase, acyl-CoA retinol acyltrunderase, folate hydrolases, protein
 - 15 phosphates (pp) 4 st, PP-1, PP-2A, PP-2Bpp-2C, deamidase, carboxyesterase, Endopeptidases, Enterokinase, Neutral endopeptidase E.C.3.4.24.11, Neutral endopeptidase, carboxypeptidases, dipeptidyl carboxypeptidase, also called peptidyl-dipeptidase A or angiotensin-converting enzyme (ACE) E.C.3.4.15.1, carboxypeptidase M, g-Glutamyl transpeptidase E.C.2.3.2.2, Carboxypeptidase P, Folate conjugase
 - 20 E.C.3.4.12.10, Dipeptidases, Glutathione dipeptidase, Membrane Gly-Leu peptidases, Zinc-stable Asp-leu dipeptidase, Enterocytic intracellular peptidases, Amino tripeptidase E.C.3.4.11.4, Amino dipeptidase E.C.3.4.13.2, Prodipeptidase, Arg-selective endoproteinase; the family of brush border hydrolases, Endopeptidase-24.11, Endopeptidase-2(meprin), Dipeptidyl peptidase IV, Membrane dipeptidase GPI,
 - 25 Glycosidases, Sucrase-isomaltase, Lactase-glycosyl-ceraminidase, Glucoamylase-maltase, Trehalase, Carbohydrase enzymes, alfa-Amylase (pancreatic), Disaccharidases (general), Lactase-phhlORIZIN hydrolase, Mammalian carbohydrases, Glucoamylase, Sucrase-Isomaltase, Lactase-glycosyl ceramidase, Enzymatic sources of ROM, Xanthine oxidase, NADPH oxidase, Amine oxidases, Aldehyde oxidase,
 - 30 Dihydroorotate dehydrogenase, Peroxidases, Human pancreatic exocrine enzymes,

Trypsinogen 1, Trypsinogen 2, Trypsinogen 3, Chymotrypsinogen, proElastase 1, proElastase 2, Protase E, Kallikreinogen, proCarboxypeptidase A1, proCarboxypeptidase A2, proCarboxypeptidase B1, proCarboxypeptidase B2, Glycosidase, Amylase, lipases, Triglyceride lipase, Collipase, Carboxyl ester hydrolase, Phospholipase A2, Nucleases, Dnase I, Ribonucleotide reductase (RNRs), Wistar rat exocrine pancreatic proteins, Label Protein IEP, A1 Amylase 1, A2 Amylase 2, Lipase, CEL Carboxyl-ester lipase, PL Prophospholipase A, T1 Trypsinogen 1, T2 Trypsinogen 2, T3 Trypsinogen 3, T4 Trypsinogen 4, C1 Chymotrypsinogen 1, C2 Chymotrypsinogen 2, PE1 Proelastase 1, PE2 Proelastase 2, PCA Procarboxypeptidase A1, PCA1 Procarboxypeptidase A2, PCB1 Procarboxypeptidase B1, PCB2 Procarboxypeptidase B2, R Ribonuclease, LS Lithostatin, Characteristics of UDPGT isoenzymes purified from rat liver, 4-nitrophenol UDPGT, 17 β -Hydroxysteroid UDDPGT, 3- α -Hydroxysteroid UDPGT, Morphine UDPGT, Billirubin UDPGT, Billirubin monoglucuronide, Phenol UDPGT, 5-Hydroxytryptamine UDPGT, Digitoxigenin monodigitoxide UDPGT, 4-Hydroxybiphenyl UDPGT, Oestrone UDPGT, Peptidases, Aminopeptidase N, Aminopeptidase A, Aminopeptidase P, Dipeptidyl peptidase IV, b-Casomorphin, Angiotensin-converting enzyme, Carboxypeptidase P Angiotensin II, Endopeptidase-24.11, Endopeptidase-24.18 Angiotensin I, Substance P (deamidated), Exopeptidase, 1. NH₂ terminus Aminopeptidase N (EC 3.4.11.2), Aminopeptidase A (EC 3.4.11.7), Aminopeptidase P (EC 3.4.11.9), Aminopeptidase W (EC 3.4.11.-), Dipeptidyl peptidase IV (EC 3.4.14.5), g-Glutamyl transpeptidase (EC 2.3.2.2), 2. COOH terminus Angiotensin-converting enzyme (EC 3.4.15.1), Carboxypeptidase P (EC 3.4.17.-), Carboxypeptidase M (EC 3.4.17.12), 3. Dipeptidase Microsomal dipeptidase (EC 3.4.13.19), Gly-Leu peptidase, Zinc stable peptidase, Endopeptidase Endopeptidase-24.11 (EC 3.4.24.11), Endopeptidase-2 (EC 3.4.24.18, PABA-peptide hydrolase, Meprin, Endopeptidase-3, Endopeptidase (EC 3.4.21.9), GST A1-1, Alpha, GST A2-2 Alpha, GST M1a-1a Mu, GST M1b-1b Mu, GST M2-2 Mu, GST M3-3 Mu, GST M4-4 Mu, GST M5-5 Mu, GST P1-1 Pi, GST T1-1 Theta, GST T2-2 Theta, Microsomal Leukotriene C₄ synthase, UGT isozymes, UGT1.1, UGT1.6, UGT1.7, UGT2.4, UGT2.7, UGT2.11, Pancreatic

enzymes, Elastase, Aminopeptidase (dipeptidyl aminopeptidase (IV), Chymotrypsin, Trypsin, Carboxypeptidase A, Methyltransferases, O-methyltransferases, N-methyltransferases, S-methyltransferases, Catechol-O-methyltransferases, MN-methyltransferase, S-sulphotransferases, Mg^{2+} -ATPase, Growth factor receptors

5 Alkaline phosphatase, ATPases, Na, K+ATPase, Ca^{2+} -ATPase, Leucine aminopeptidase, K^+ channel. The metabolic enzymes of the present invention and the like may be immobilized onto the sensing surface through the use of known immobilization techniques as is appreciated by those skilled in the art.

Suitable transport proteins for estimating an "excretion" (as well as an

10 efflux, absorption, distribution) parameter of a drug candidate include Glucose a. Na^+ -glucose cotransp GLUT 1, b. Facilitative transporter, GLUT 1-5, GLUT-2, Neutral amino acid transporter, Na^+ -independent system L amino acid transporter, cationic amino acid, Y^+ cationic L-amino acid transporter, Dipeptides H^+ cotransport, Nucleosides Na^+ -dependent and facilitative, Taurine Na^+ and Cl^- dependent, Bile acids

15 Na^+ /bile acid cotransporter, Na^+ -independent bile acid transporter, ABC transporters, Prostaglandins facilitative transporter, Na^+/H^+ exchanger Antiporter, Phosphate Na^+/Pi cotransporter, Sulfate Na^+ -cotransporter, Transporters for neurotransmitters, Norepinephrine Na^+/Cl^- cotransporter, Dopamine Na^+/Cl^- cotransporter, Serotonine Na^+/Cl^- cotransporter, GABA, GAT-1, Na^+/Cl^- dependent, Glycine Na^+/Cl^- dependent,

20 Glutamate Na^+ cotransporter, K^+/OH^- counter-transport, ABC transporters, P-glycoprotein (MDR1, MDR 2 OR MDR 3), Cl^- channel (CFTR), Antigenic peptides, TAP1 and TAP2 heterodimer, Lung resistance protein (LRP), Multidrug resistance protein 1 (MRP1), Multidrug resistance protein 2 (MRP2 or cMOAT), Multidrug resistance protein 3 (MRP3), Multidrug resistance protein 4 (MRP4), Multidrug

25 resistance protein 5 (MRP5), Multidrug resistance protein 6 (MRP6), mrp (mouse), EBCR (rabbit), *C. elegans* mrp1 (nematode), *C. elegans* mrp2 (nematode), MRP6 (human), YCF1 (yeast), AtMRP1 (Arabidopsis), SUR1 (human), sur2 (rat, mouse), YOR1/YRS1 (yeast), LtpgpA (leishmania), Hepatic amino acid transport system, Neutral amino acid transporters, MeAIB, Dicarboxylic amino acids, Neutral amino

30 acids (branched), b-amino acids, Long chain fatty acids, Monoglycerides, L-

5 Lysophosphatidylcholine, Transport proteins for bilirubin, Bilitranslocase (BTL),
 Organic anion binding protein (OABP), BSP/bilirubin binding protein, Signal receptor
 and transduction-hydrolases, ATPases, Na⁺dependent/independent bile acid transport,
 Bilirubin/BSP carrier (Cl⁻-dependent), SO₄/OH⁻ exchanger Cl⁻channel Na⁺/H⁺ exchanger,
 10 Na⁺/HCO₃⁻ cotransport, GSH-, GSSH-, GS conjugate carrier, SO₄²⁻/HCO₃⁻ exchanger, Na⁺-
 dependent amino acid transport, Dipolar amino acid transporter, Basic amino acids,
 Cystine, Imino acids Cl⁻, b-Amino acids Cl⁻, XAG Acidic amino acids K⁺, A Dipolar a-
 amino acids, three-and four-carbon dipolar amino acids, L Bulky, hydrophobic, dipolar
 amino acids, γ+ Basic amino acids, folate transporter, cbl transport proteins, Na⁺-K⁺ -
 15 ATPase, Bile acid transporter (BAT), protein kinase C, Na⁺/I⁻ symporter (NIS); Bile salt
 export pump(BSEP, cBAT, SPGP); Intestinal bile acid transporters; Purine selective
 Na⁺-dependent nucleoside transporter (hSPNTI); Pyrimidine selective Na⁺-dependent
 nucleoside transporter (cNTI); Mitoxantrone transporter (MXR1 and MXR2); Intestinal
 oligopeptide transporter(PepT1); Renal oligopeptide transporter(PepT2); Breast cancer
 20 resistance protein (BCRP). The transport/efflux proteins of the present invention and
 the like may be immobilized onto the sensing surface through the use of known
 immobilization techniques as is appreciated by those skilled in the art.

As stated above, it has been discovered that the pharmacokinetic
 parameters of absorption, distribution, metabolism, and excretion (ADME) of a drug
 25 candidate may be determined from the binding interactions between the drug candidate
 and appropriate sensing surface-bound biomolecules (*e.g.*, from different liposomes,
 plasma proteins, CYP 450 enzymes, other metabolic enzymes, and/or transport/efflux
 proteins, such as those as identified above) of a biosensor. That is, the binding
 interactions between a drug candidate and one or more sensing surface-bound
 30 biomolecules selected from liposomes, plasma proteins, CYP 450 enzymes, other
 metabolic enzymes, and transport/efflux proteins, may be measured via the BIACORE
 instrument to determine a "binding interaction parameter" of the drug candidate. The
 estimated binding interaction parameter may then, in turn, be compared against a
 predetermined drug correlation graph to estimate one or more absorption, distribution,
 35 metabolism, and excretion "pharmacokinetic parameters" of the drug candidate.

Moreover, the pharmacokinetic parameter of the drug candidate may be, for example, selected from the group consisting of volume of distribution; total clearance; protein binding; tissue binding; metabolic clearance; renal clearance; hepatic clearance; biliary clearance; intestinal absorption; bioavailability; relative bioavailability; intrinsic
 5 clearance; mean residence time; maximum rate of metabolism; Michaelis-Menten constant; partitioning coefficients between tissues and blood (or plasma) such as those partitioning coefficients associated with the blood brain barrier, blood placenta barrier, blood human milk partitioning, blood adipose tissue partitioning, and blood muscle partitioning; fraction excreted unchanged in urine; fraction of drug systemically
 10 converted to metabolites; elimination rate constant; half-life; and secretion clearance; any one of which may be determined from appropriate sensorgrams of the selected drug-biomolecule interaction (alternatively, they may be determined by extracting score vectors with principal component analysis from digitalized interaction profiles, as well as other known multivariate methods).

15 For example, to estimate an apparent equilibrium constant between a drug candidate and selected sensing surface-bound biomolecules, the following procedure may be employed. First, a concentration series (*e.g.*, 0, 20, 50, 100, 500, and 1,000 μM) of the drug candidate may be prepared, and sequentially injected into a biosensor having a sensor chip operatively associated therewith, wherein the sensor chip
 20 has a reference sensing surface and at least one sensing surface with surface-bound biomolecules. The relative responses at steady-state binding levels for each drug concentration level may then be measured. Because of bulk-refractive index contributions from solvent additives in the biosensor's running buffer, a correction factor may be calculated (via known calibration procedures) and applied to give
 25 corrected relative responses. The corrected relative responses for each drug concentration may then be mathematically evaluated as is appreciated by those skilled in the art to estimate the apparent equilibrium constant.

More specifically, the apparent equilibrium constant (K_D) between the drug candidate and the selected sensing surface-bound biomolecules may be calculated

by fitting the measured equilibrium (R_{eq}) data and known drug concentration (C) data to Equation (1):

$$R_{eq} = C * R_{max} / (C + KD) + \text{offset} \quad (1)$$

wherein R_{max} is the maximum binding capacity of the sensing surface. Alternatively, the apparent equilibrium constant (KD) may be calculated from as little as two concentrations (*i.e.*, $C1$ and $C2$) of the drug candidate by use of Equation (2):

$$KD = (R_{eqC1}/C1 - R_{eqC2}/C2) / (R_{eqC2} - R_{eqC1}) \quad (2)$$

The estimated apparent equilibrium constant (KD) or binding level at a specific molar drug concentration may then, in turn, be compared against a predetermined drug correlation graph to estimate one or more absorption, distribution, metabolism, and excretion parameters of the drug candidate.

In the context of the present invention, a predetermined drug correlation graph refers to a mathematical expression or function that has been developed from binding interaction data associated with known drug compounds. For example, a correlation graph may be constructed wherein known binding interaction parameters (*e.g.*, apparent equilibrium constants and/or binding levels at specific molar drug concentrations) for known compounds are plotted along the abscissa (*i.e.*, the x-axis) and corresponding measured binding interaction parameters obtained via the biosensor are plotted along the ordinate (*i.e.*, the y-axis), or vice versa. Stated somewhat differently, the correlation plot is a graphical representation of a mathematical expression that correlates known and measured affinity data.

By comparing the estimated binding interaction parameter obtained from the selected drug-biomolecule interaction against the mathematical expression (*i.e.*, correlation graph) correlated from known and measured affinity data, it has been surprisingly discovered that an estimate of a pharmacokinetic ADME parameter related to the drug candidate may be accurately predicted. In addition, by immobilizing a selected combination of at least two different liposomes, plasma proteins, CYP 450 enzymes, other metabolic enzymes, and/or transport/efflux proteins, onto one or more

discrete sensing surfaces of a sensor chip, it has been further surprisingly discovered that at least two pharmacokinetic parameters associated with a drug candidate may be readily determined. Moreover, by immobilizing a selected combination of different liposomes, plasma proteins, CYP 450 enzymes, other metabolic enzymes, and/or
5 transport/efflux proteins, onto the sensing surfaces of biosensor, such as in a predetermined line of different spots of biomolecules, an ADME pattern or profile, such as a drug candidate characterization matrix, may be readily developed. Such ADME profiles are of great utility for purposes of drug screening.

In addition to determining one or more pharmacokinetic parameters by
10 monitoring the refractive index changes of a biosensor as disclosed above, the solubility of a drug candidate may also be simultaneously determined (together with the one or more pharmacokinetic parameters) by monitoring the minimum, maximum or centroid of the drug-biomolecule interaction signal (as is disclosed, for example, in PCT Publication No. WO 97/09618, which is incorporated herein by reference in its
15 entirety). More specifically, the solubility of the drug candidate may be estimated/identified from irregularities, as well as from reflectance minimum (R_{min}) and dip-shape data, shown in the sensorgram of the drug-biomolecule interaction. In general, the concentration at which precipitation occurs is referred to as the solubility limit; this property may be important to measure because it may indicate that the drug
20 candidate has an affinity greater than otherwise indicated.

Solubility problems of the drug candidate may be detected because insoluble particles (*i.e.*, precipitates) tend to bind to sensing surface-bound biomolecules, thereby causing the sensing surface to be non-homogenous. (A homogenous surface has the same surface concentration throughout, whereas a non-
25 homogenous surface has concentration disruptions.) A non-homogenous sensing surface actually measures several refractive indices, which are all averaged together in the biosensor's detector. These multiple measurements from a single, but non-homogenous sensing surface, tend to result in an increase in reflectance minimum and a broadening of the dip associated therewith.

An illustration of how sensorgram irregularities may be used to identify solubility problems (*i.e.*, the presence of precipitates) is shown in Figures 1A-C. More specifically, the steady-state binding levels associated with a selected drug-biomolecule interaction is generally shown in Figure 1A. By enlarging or "zooming" in on the top area of the sensorgram, which is reflective of the steady-state binding, the sensorgram irregularities become more apparent as is shown in corresponding Figure 1B. The sensorgram irregularities become even more apparent by using reference subtracted data (*i.e.*, bulk-refractive index effects have been eliminated) as is shown in corresponding Figure 1C.

An illustration of how reflectance minimum (R_{\min}) and dip-shape data may be used to identify solubility problems (*i.e.*, the presence of precipitates) is shown in Figures 2A-C. More specifically, a homogeneous surface has the same surface concentration throughout, and will thus result in a single "dip" with respect to the reflected light intensity as shown in Figure 2A. When precipitates bind to the sensing surface, the sensing surface becomes non-homogenous which tends to result in an increase in reflectance minimum and a broadening of the dip associated therewith as is shown in Figures 2B and 2C, respectively.

Based on the foregoing methods for assaying a drug candidate, researchers may now simultaneously measure several different pharmacokinetic parameters of the drug candidate, as well as gauge the drug candidate's solubility, by using a single analytical instrument. The present invention simplifies and improves the rate of drug discovery and development because important pharmacokinetic data may now be readily obtained at a relatively early stage of the process.

In addition to the foregoing methods, the present invention is also directed to apparatuses adapted to carrying out such methods. More specifically, the apparatuses of the present invention comprise a biosensor having a sensing surface associated therewith, and a computer system that facilitates the implementation of the steps associated with the methods disclosed herein. The computer includes a computer memory containing a data structure useful for assaying a drug candidate; the data structure comprises binding interaction data associated with known drug compounds

such that the data structure may be used to determine an estimate of at least pharmacokinetic parameter of the drug candidate.

The aspect of the present invention relating to a computer memory containing a data structure useful for assaying a drug candidate may be more fully illustrated in the context of a high-level computer block diagram as is depicted in Figure 3. As shown, such a computer system 300 contains one or more central processing units (CPUs) 310, input/output devices 320, and the computer memory containing a data structure useful for assaying a drug candidate(memory) 330. Among the input/output devices is a storage device 321, such as a hard disk drive, and a computer-readable media drive 322, which may be used to install software products, where the software products are provided on a computer-readable medium, such as a CD-ROM. The input/output devices also include a network connection 323, through which the computer system 300 may communicate with other connected computer systems, such as networks. The input/output devices may also contain a display 324 and a data input device 325.

The memory 330 preferably contains an operating system 331, such as MICROSOFT WINDOWS, for providing to other programs access to resources of the computer system. The memory 330 preferably further contains software 332. While the computer memory containing a data structure useful for assaying a drug candidate is preferably implemented on a computer system configured as described above, those skilled in the art will recognize that it may also be implemented on computer systems having different configurations.

In a related aspect, the present invention is also directed to a generated data signal conveying a data structure useful for assaying a drug candidate. As above, the data structure comprises binding interaction data associated with known drug compounds, such that the data structure may be used to determine an estimate of at least pharmacokinetic parameter of the drug candidate.

In still a further embodiment, a sensor surface adapted for use with a biosensor is disclosed. The sensor surface has a hydrogel matrix coating coupled to the top surface of the sensor surface, wherein the hydrogel matrix coating has a plurality of

functional groups, and at least two different types of liposomes are bonded to the plurality of functional groups at discrete and noncontiguous locations on the hydrogel matrix coating of the sensor surface (such as disclosed in Example 1). In the context of the BIACORE instrument as described above, the sensor surface is preferably in the form a sensor chip, wherein the sensor chip has a free electron metal interposed between the hydrogel matrix and the top surface of the sensor chip. Suitable free electron metals in this regard include copper, silver, aluminum and gold.

The sensor surface may have a lipophilic substance interposed between the different types of liposomes and the plurality of functional groups, wherein the lipophilic substance is covalently bonded to the plurality of functional groups. Representative lipophilic substances comprise an alkyl chain having from 12 to 24 carbon atoms, such as stearylamine. Similarly, representative liposomes include 1,2-dimyristol-*sn*-glycero-3-phosphocholine (DMPC) and 1-palmitoyl-2-oleoyl-*sn*-glycero-3-phosphocholine (POPC). The sensor surfaces of this invention may also have non-liposome biomolecules associated therewith. For example, human serum albumin, CYP 450 enzyme, a metabolic enzyme, and/or a transport protein may be bonded to the plurality of may be bonded to the plurality of functional groups at discrete and noncontiguous locations on the hydrogel matrix coating of the sensor surface.

For purposes of illustration and not limitation, the following examples more specifically disclose various aspects of the present invention.

EXAMPLES

EXAMPLE 1

SIMULTANEOUS MEASUREMENT OF SOLUBILITY, PLASMA PROTEIN BINDING, LIPOPHILICITY AND INTESTINAL ABSORPTION FOR THREE DRUG CANDIDATES A-C

This example discloses how combined information from each of four different flow-cells of a biosensor may be represented in a characterization matrix where each of the three drug candidates A-C illustrates a quality pattern (*i.e.*, HSA% Bound, Predicted Lipophilicity, Solubility, and Predicted FA%) which is useful for the selection of lead drug compounds.

Preparation of Sensor Chip

Three of the four discrete sensing surfaces of a CM5 Sensor Chip (Biacore AB, Uppsala, Sweden) were modified such that the CM5 Sensor Chip had surface-bound biomolecules as depicted below in Table 1.

5

TABLE 1

SURFACE-BOUND BIOMOLECULES OF CM5 SENSOR CHIP

Surface/Cell No.	Type of Flow-Cell	Surface Modification
FC1	Ref-1	Unmodified carboxymethyl dextran (CM5)
FC2	Target-1	Human Serum albumin - HSA (9-12 kRU)
FC3	Target-2	DMPC-liposomes (5-7 kRU) captured on stearylamine
FC4	Target-3	POPC-liposomes (5-7 kRU) captured on stearylamine

More specifically, and to achieve covalent attachment of stearylamine to surface/cell nos. 3 and 4, the following procedure was employed. A CM5 Sensor Chip was first inserted into a BIACORE 3000 biosensor (Biacore AB, Uppsala, Sweden). The flow of running buffer (isotonic phosphate buffer pH 7.0 (9.6 g $\text{Na}_2\text{HPO}_4 \cdot 2\text{H}_2\text{O}$, 1.7 g KH_2PO_4 , 4.1 g NaCl to 1 liter)) with 2% dimethylsulfoxide (DMSO) was directed to flow-cells 3 and 4 by using appropriate software commands. A mixture of 0.2 M N-ethyl-N'-(3-diethylaminopropyl)-carbodiimide and 50 mM N-hydroxysuccinimide in water was then injected into the biosensor so as to flow over flow-cells 3 and 4 for a period of 10 minutes. The CM5 Sensor Chip was then washed with running buffer, removed from the biosensor, and placed in a petri dish on top of a few layers of tissue paper that had been moistened with ethanol. The exposed sensing surfaces were then treated with 30 μl of a 10 mM stearylamine solution in 99% ethanol. After another 45 minute period, the CM5 sensor chip was gently washed with ethanol and then with water. Next, the exposed sensing surfaces were further treated with 50 μl of a 1M ethanolamine solution at pH 8.5. Finally, and after another 10 minute period, the sensing surfaces were washed with HBS-buffer.

Following covalent attachment of stearylamine to surface/cell nos. 3 and 4 - and after the CM5 Sensor Chip that had just been washed with HBS-buffer was blown dry with nitrogen and reinserted into the biosensor - human serum albumin (HSA), DMPC-liposomes, and POPC-liposomes (available from Sigma) were then captured onto surface/cell nos. 2, 3 and 4, respectively. The following procedures were employed to capture these biomolecules.

To capture human serum albumin the flow was initially directed to flow-cell 2 by using appropriate software commands. The sensing surface of flow-cell 2 was then activated for a period of 7 minutes with EDC/NHS. Next, human serum albumin was injected at 15 µg/ml in 10mM acetate buffer pH 5.2 for a period of 7 minutes; then 1M ethanolamine pH 8.5 was injected for another period of 7 minutes. This procedure resulted in the immobilization of 9-12 kRU of human serum albumin to the sensing surface of surface/cell no. 2.

To capture the two different liposomes the flow was first directed to flow-cell 3, and 0.5-1 mM DMPC liposome was injected until 5-7 kRU of the DMPC liposome had been captured. The flow was then directed to flow-cell 4, and 0.5-1 mM POPC liposome was injected until 5-7 kRU of the POPC liposome had been captured. Finally, the flow-system of the biosensor was washed with 100mM NaOH, and the autoinjector tubing was washed with 0.5% SDS and 50 nM glycine pH 9.5 to remove trace lipids.

Calibration Procedure

In order to improve data quality by reducing bulk-refractive index contributions from the DMSO solvent additive, a calibration procedure was employed. (Note the calibration procedure required varying concentrations of DMSO.) Because the running buffer contained 2% DMSO, a series of 8 to 10 calibration solutions were made that had varying DMSO concentrations ranging from 1.5% to 3.0%. Each calibration solution was sequentially injected over each of the four sensing surfaces by using the serial injection mode of the biosensor, and the respective steady-state binding levels were measured. The relative responses of the reference sensing surface of flow-

cell 1 (FC1 having unmodified CM5 as a reference surface) were then subtracted from the corresponding calibration responses of the target sensing surfaces of flow-cells 2, 3, and 4, respectively, and plotted as functions of the relative responses. From these plots, calibration functions were calculated for each of the respective sensing surfaces.

5 The calibration functions were then used to calculate appropriate correction factors for the samples containing the three drug candidates A-C. That is, for the samples containing the drug candidates, the relative responses of the steady state binding levels of the reference sensing surface (*i.e.*, signals from flow-cell 1) were measured and the respective calibration functions were used to calculate correction
10 factors appropriate for each drug candidate. The correction factors were then applied (*i.e.*, subtracted) to the differences between the reference responses and the sample responses to give a responses where the bulk-refractive index contributions from the DMSO solvent additive had been eliminated.

Corrected Relative Responses of Drug Candidates A-C

15 The modified CM5 Sensor Chip (having surface-bound biomolecules as depicted above in Table 1) was used to measure the binding interaction of the three drug candidates A-C. Concentration series (*i.e.*, 0, 20, 50, 100, 500, and 1,000, μM) of each drug candidate were prepared, and sequentially injected in serial mode into the biosensor such that four sensorgrams corresponding to each flow-cell were developed.
20 Corrected relative responses at steady-state binding levels for each drug concentration were determined for each of the target sensing surfaces/flow-cells; such data is presented below in Tables 2-4.

TABLE 2

DRUG CONCENTRATION DATA FOR HUMAN SERUM ALBUMIN

Candidate	10 μM	50 μM	100 μM	500 μM	1,000 μM
A	50	110	130	140	145
B	0	0	0	10	50
C	10	50	100	110	115

TABLE 3
DRUG CONCENTRATION DATA FOR POPC LIPOSOME

Candidate	10 μ M	50 μ M	100 μ M	500 μ M	1,000 μ M
A	10	300	400	450	500
B	0	0	40	60	70
C	0	0	0	0	10

5

TABLE 4
DRUG CONCENTRATION DATA FOR DMPC LIPOSOME

Candidate	10 μ M	50 μ M	100 μ M	500 μ M	1,000 μ M
A	0	50	70	90	95
B	0	10	20	50	55
C	0	0	0	0	0

10 Analysis of Data

The corrected relative responses for each drug concentration as depicted above in Tables 2-4 were, among other things, adjusted for differences in molecular weight of the drug candidates, as well as transformed to increase correlation with other properties. For example, for each target sensing surface/flow-cell, an apparent equilibrium constant (KD) was calculated by fitting the measured equilibrium (R_{eq}) data and known drug concentration (C) data to Equation (3):

$$R_{eq} = C \cdot R_{max} / (C + KD) + \text{offset} \quad (3)$$

20 wherein R_{max} is the maximum binding capacity of the sensing surface. In the fitting procedure, R_{max} , KD, and the offset were calculated.

Alternatively, the apparent equilibrium constant (KD) may have been calculated from two concentrations of the drug candidate, C1 and C2, by use of Equation (4):

25

$$KD = (R_{eqC1}/C1 - R_{eqC2}/C2)/(R_{eqC2} - R_{eqC1}) \quad (4)$$

In either case, the KD values obtained via the biosensor were subsequently correlated with known KD values calculated from reported human serum albumin binding percentages (*i.e.*, reported binding percentages that have been reported for known drug compounds), thereby enabling a prediction of the degree of protein binding for each of the three drug candidates A-C. In other words, a correlation graph as shown in Figure 4 was constructed having known KD values for known compounds plotted along the abscissa (*i.e.*, the x-axis) and corresponding measured KD values obtained via the biosensor plotted along the ordinate (*i.e.*, the y-axis). This correlation graph was then used to predict the degree of plasma protein binding for each of the three drug candidates A-C (see below).

In addition, for each target sensing surface/flow-cell, a molecular weight adjusted response at a single concentration level was used as a threshold value for ranking the drug candidates A-C. More specifically, a correlation graph as shown in Figure 5 was constructed, wherein respective binding levels at 100 μ M ($R_{100 \mu M}$) divided by the known drug compounds' molecular weight were plotted along the abscissa (*i.e.*, the x-axis) and corresponding human serum albumin binding percentage, as measured by equilibrium dialysis, were plotted along the ordinate (*i.e.*, the y-axis). This correlation graph was then used to discriminate between strong and weak human serum albumin binders for each of the three drug candidates A-C.

Furthermore, for each target sensing surface/flow-cell, reference subtracted sensorgram traces (*e.g.*, FC2 minus FC1) were developed to detect the presence of precipitates (caused by low solubility). More specifically, reference subtracted sensorgram traces for each of the three drug candidates A-C, as shown in Figure 6, were constructed. (Note that when precipitates form and then bind/associate with the sensing surface, the response signal will decrease in a non-continuous way.) By analyzing the concentration series for each of the three drug candidates A-C (as shown in Figure 6), it was determined that drug candidate A precipitated and thus had relatively low solubility, whereas drug candidates B-C did not precipitate and thus had relatively high solubilities.

Finally, and as shown below in Table 5, the combined information from each target sensing surface/flow cell was tabulated into a reduced characterization matrix where each drug candidate A-C received a quality pattern that was useful for selection of the lead drug compounds. (Note that the predicted lipophilicity and fraction absorbed are prophetic.)

TABLE 5
CHARACTERIZATION MATRIX FOR DRUG CANDIDATES A-C

Candidate	HSA % Bound	Pred. Lipophilicity	Solubility	Predicted FA%
A	99.9	4	<1 μ M	>97
B	<90	2	OK	>90
C	>90	-2	OK	<50

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Based on the forgoing characterization matrix, it was determined that drug candidate B was the preferred compound due to its low plasma protein binding level (HSA %bound <90), medium lipophilicity (predicted lipophilicity = 2), no identified solubility problems (solubility = OK), and acceptable intestinal absorption (predicted fraction absorbed >90).

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EXAMPLE 2

DEMONSTRATED CORRELATION BETWEEN BIOSENSOR DATA AND FRACTION ABSORBED IN HUMANS (FA%)

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This example discloses a correlation between biosensor data obtained from the BIACORE instrument (*i.e.*, BIACORE 3000) and known data for the fraction absorbed in humans (FA%) for a number of different drugs, wherein the correlation graph is useful for drug candidate absorption predictions. More specifically, a correlation graph as shown in Figure 7 was constructed having known fraction absorbed in humans (FA%) plotted along the ordinate (*i.e.*, the y-axis) and corresponding calibrated (*i.e.*, reference subtracted) steady state binding levels for each drug at 500

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5 μM plotted along the abscissa in a 10-logarithm scale (*i.e.*, the x-axis). In this example, the sensing surfaces of the target flow-cells each had 6,000 RU of POPC-GM3 ganglioside (available from Sigma) captured on stearylamine tiles. (Note that an unmodified sensing surface of a CM5 Sensor Chip was used as the reference.) As shown in Figure 7, there is high degree of correlation between the biosensor data and the known fraction absorbed in humans (FA%) data, as is evidenced by the mathematical expression/function that has been fitted to the various data points.

10 Moreover, the correlation graph also shows a classification of the various substances, wherein absorption of the substances using the passively transported trans-cellular route through the intestine are depicted by h = high, m = medium, and l = low, and wherein absorption using active transport is depicted by h-act = high, and wherein absorption of substances with molecular weights <200 using the para-cellular route is depicted by h-para = high and m-para = medium, respectively. (Note the Sulfasalazine is a pre-drug which rapidly decomposes in the intestine, which may explain its outlier properties; Piroxicam has a very low solubility $<<<<500 \mu\text{M}$, which may explain its lack of correlation.)

20 While the present invention has been described in the context of the embodiments illustrated and described herein, the invention may be embodied in other specific ways or in other specific forms without departing from its spirit or essential characteristics. Therefore, the described embodiments are to be considered in all respects as illustrative and not restrictive. The scope of the invention is, therefore, indicated by the appended claims rather than by the foregoing description, and all changes that come within the meaning and range of equivalency of the claims are to be

25 embraced within their scope.